

In []:

```

import numpy as np
import pandas as pd

import matplotlib.pyplot as plt
%matplotlib inline

from keras.models import Sequential
from keras.layers import Dense, Dropout, Lambda, Flatten
from keras.optimizers import Adam, RMSprop
from sklearn.model_selection import train_test_split
from keras import backend as K
from tensorflow.keras.preprocessing.image import ImageDataGenerator

from subprocess import check_output
print(check_output(["ls", "input"]).decode("utf8"))

```

test.csv
train.csv

In [2]:

```

# create the training & test sets, skipping the header row with [1:]
train = pd.read_csv("input/train.csv")
print(train.shape)
train.head()

```

(42000, 785)

Out[2]:

	label	pixel0	pixel1	pixel2	pixel3	pixel4	pixel5	pixel6	pixel7	pixel8	...	pixel7
0	1	0	0	0	0	0	0	0	0	0	...	
1	0	0	0	0	0	0	0	0	0	0	...	
2	1	0	0	0	0	0	0	0	0	0	...	
3	4	0	0	0	0	0	0	0	0	0	...	
4	0	0	0	0	0	0	0	0	0	0	...	

5 rows × 785 columns



In [3]:

```

test = pd.read_csv("input/test.csv")
print(test.shape)
test.head()

```

(28000, 784)

Out[3]:

	pixel0	pixel1	pixel2	pixel3	pixel4	pixel5	pixel6	pixel7	pixel8	pixel9	...	pixel
0	0	0	0	0	0	0	0	0	0	0	...	
1	0	0	0	0	0	0	0	0	0	0	...	
2	0	0	0	0	0	0	0	0	0	0	...	
3	0	0	0	0	0	0	0	0	0	0	...	
4	0	0	0	0	0	0	0	0	0	0	...	

5 rows × 784 columns



In [4]: `X_train = (train.iloc[:,1:].values).astype('float32') # all pixel values`
`y_train = train.iloc[:,0].values.astype('int32') # only labels i.e targets digit`
`X_test = test.values.astype('float32')`

In [5]: `X_train`

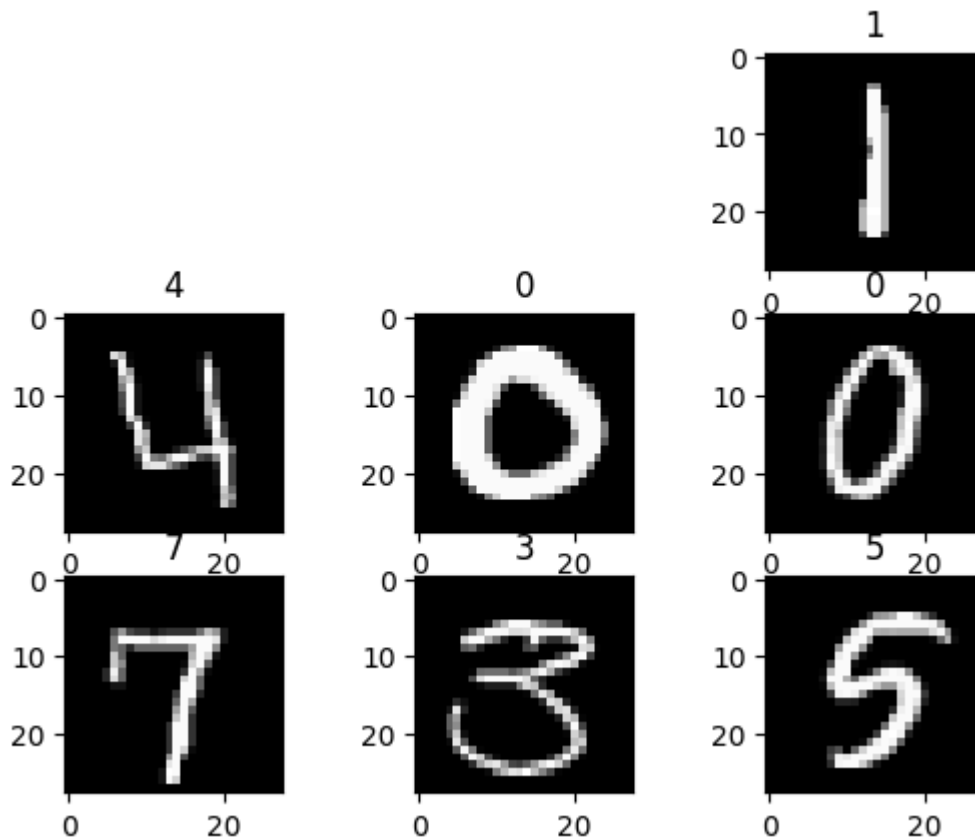
Out[5]: `array([[0., 0., 0., ..., 0., 0., 0.],`
 `[0., 0., 0., ..., 0., 0., 0.],`
 `[0., 0., 0., ..., 0., 0., 0.],`
 `...,`
 `[0., 0., 0., ..., 0., 0., 0.],`
 `[0., 0., 0., ..., 0., 0., 0.],`
 `[0., 0., 0., ..., 0., 0., 0.]], shape=(42000, 784), dtype=float32)`

In [6]: `y_train`

Out[6]: `array([1, 0, 1, ..., 7, 6, 9], shape=(42000,), dtype=int32)`

In [8]: `## Data Visualization`
`#Convert train dataset to (num_images, img_rows, img_cols) format`
`X_train = X_train.reshape(X_train.shape[0], 28, 28)`

`for i in range(2, 9):`
 `plt.subplot(330 + (i+1))`
 `plt.imshow(X_train[i], cmap=plt.get_cmap('gray'))`
 `plt.title(y_train[i]);`



```
In [9]: #expand 1 more dimation as 1 for colour channel gray
X_train = X_train.reshape(X_train.shape[0], 28, 28,1)
X_train.shape
```

Out[9]: (42000, 28, 28, 1)

```
In [10]: X_test = X_test.reshape(X_test.shape[0], 28, 28,1)
X_test.shape
```

Out[10]: (28000, 28, 28, 1)

```
In [11]: #Preprocessing the digit images
#Feature Standardization
mean_px = X_train.mean().astype(np.float32)
std_px = X_train.std().astype(np.float32)

def standardize(x):
    return (x-mean_px)/std_px
```

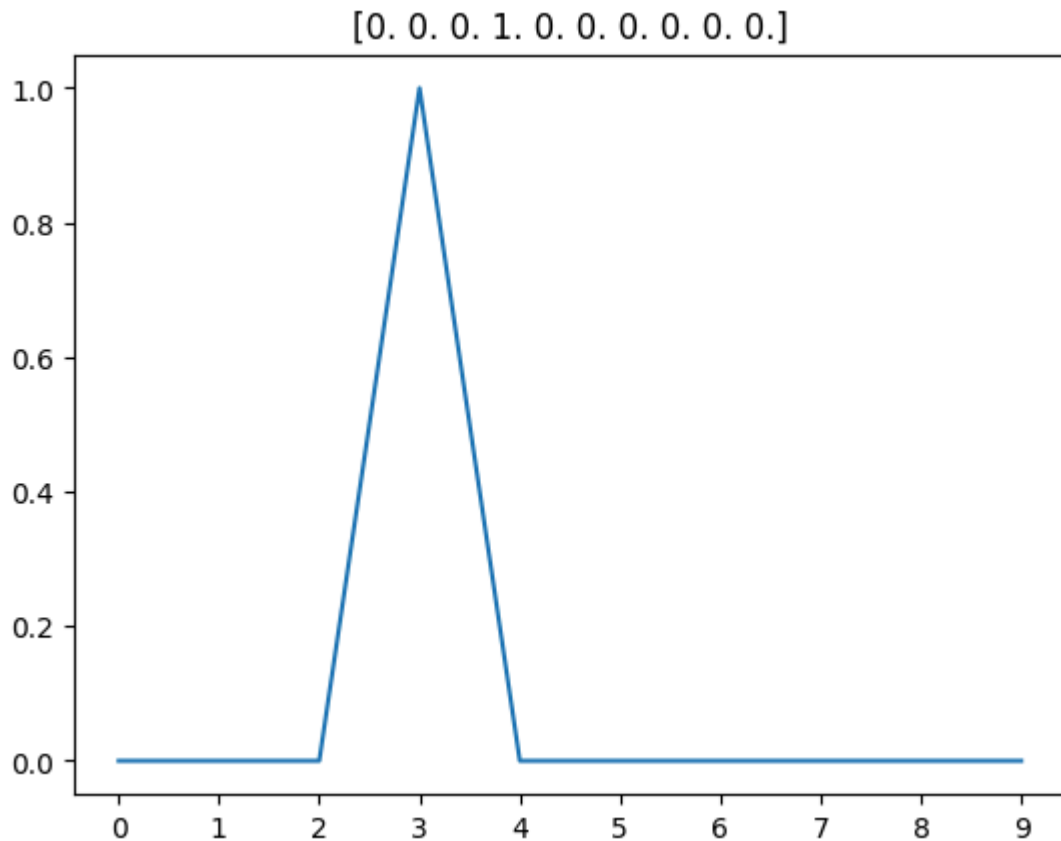
```
In [12]: """One Hot encoding of labels.
A one-hot vector is a vector which is 0 in most dimensions, and 1 in a single di
For example, 3 would be [0,0,0,1,0,0,0,0,0,0]."""
from tensorflow.keras.utils import to_categorical

y_train= to_categorical(y_train)
num_classes = y_train.shape[1]
num_classes
```

Out[12]: 10

```
In [13]: plt.title(y_train[9])
plt.plot(y_train[9])
```

```
plt.xticks(range(10));
```



```
In [14]: #Designing Neural Network Architecture
# fix random seed for reproducibility
seed = 43
np.random.seed(seed)
```

```
In [15]: #Linear Model
from keras.models import Sequential
from tensorflow.keras.layers import Lambda, Dense, Flatten, Dropout

from keras.callbacks import EarlyStopping
from keras.layers import BatchNormalization, Convolution2D, MaxPooling2D
```

```
In [16]: model= Sequential()
model.add(Lambda(standardize,input_shape=(28,28,1)))
model.add(Flatten())
model.add(Dense(10, activation='softmax'))
print("input shape ",model.input_shape)
print("output shape ",model.output_shape)
```

c:\Users\LENOVO\AppData\Local\Programs\Python\Python310\lib\site-packages\keras\src\layers\core\lambda_layer.py:65: UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When using Sequential models, prefer using an `Input(shape)` object as the first layer in the model instead.

```
super().__init__(*kwargs)
```

WARNING:tensorflow:From c:\Users\LENOVO\AppData\Local\Programs\Python\Python310\lib\site-packages\keras\src\backend\tensorflow\core.py:232: The name tf.placeholder is deprecated. Please use tf.compat.v1.placeholder instead.

```
input shape (None, 28, 28, 1)
output shape (None, 10)
```

```
In [17]: from tensorflow.keras.optimizers import RMSprop
```

```
model.compile(
    optimizer=RMSprop(learning_rate=0.001),
    loss='categorical_crossentropy',
    metrics=['accuracy']
)
```

```
In [18]: from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

```
gen = ImageDataGenerator()
```

```
In [19]: ## Cross Validation
```

```
from sklearn.model_selection import train_test_split
X = X_train
y = y_train
X_train, X_val, y_train, y_val = train_test_split(X_train, y_train, test_size=0.
batches = gen.flow(X_train, y_train, batch_size=64)
val_batches=gen.flow(X_val, y_val, batch_size=64)
```

```
In [20]: history = model.fit(
    batches,
    steps_per_epoch=len(batches),
    epochs=3,
    validation_data=val_batches,
    validation_steps=len(val_batches)
)
```

Epoch 1/3

591/591 ————— 4s 5ms/step - accuracy: 0.8641 - loss: 0.4535 - val_ accuracy: 0.9005 - val_loss: 0.3330

Epoch 2/3

591/591 ————— 3s 4ms/step - accuracy: 0.9123 - loss: 0.3037 - val_ accuracy: 0.9036 - val_loss: 0.3205

Epoch 3/3

591/591 ————— 2s 3ms/step - accuracy: 0.9194 - loss: 0.2869 - val_ accuracy: 0.9164 - val_loss: 0.2973

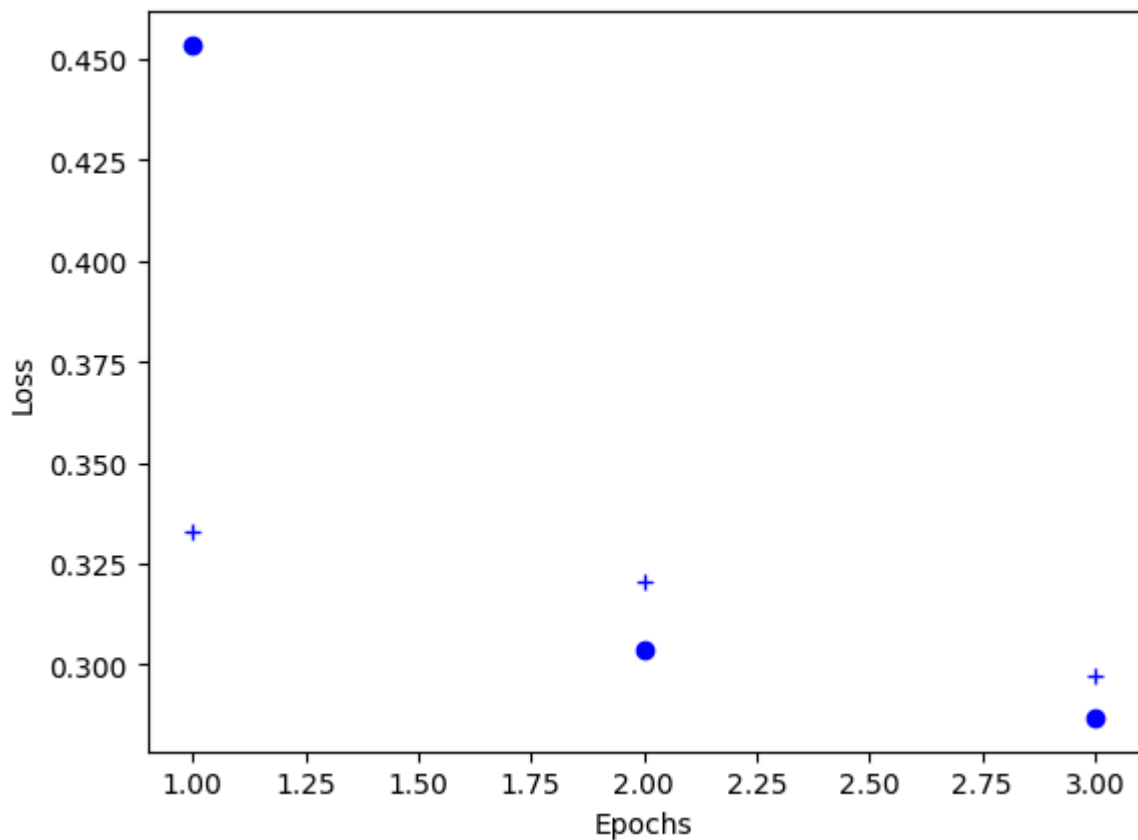
```
In [21]: history_dict = history.history
history_dict.keys()
```

```
Out[21]: dict_keys(['accuracy', 'loss', 'val_accuracy', 'val_loss'])
```

```
In [22]: import matplotlib.pyplot as plt
%matplotlib inline
loss_values = history_dict['loss']
val_loss_values = history_dict['val_loss']
epochs = range(1, len(loss_values) + 1)

# "bo" is for "blue dot"
plt.plot(epochs, loss_values, 'bo')
# b+ is for "blue crosses"
plt.plot(epochs, val_loss_values, 'b+')
plt.xlabel('Epochs')
plt.ylabel('Loss')

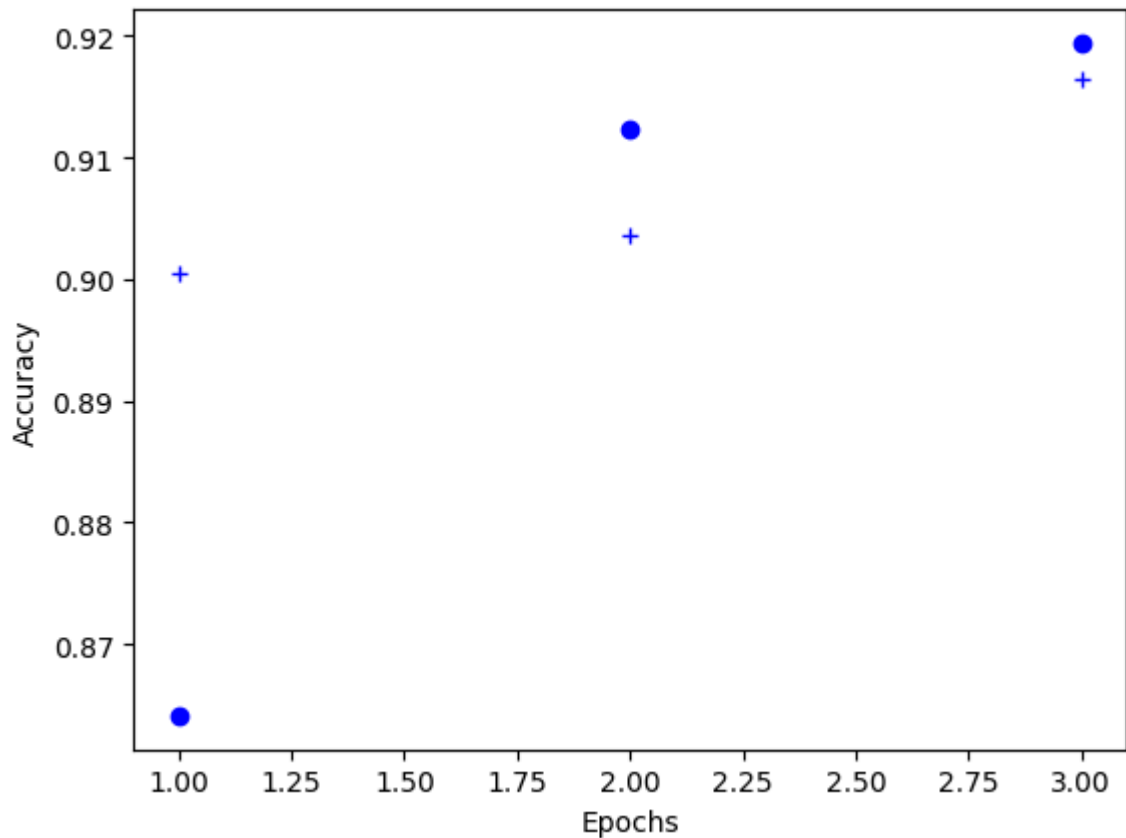
plt.show()
```



```
In [23]: plt.clf() # clear figure
acc_values = history_dict['accuracy']
val_acc_values = history_dict['val_accuracy']

plt.plot(epochs, acc_values, 'bo')
plt.plot(epochs, val_acc_values, 'b+')
plt.xlabel('Epochs')
plt.ylabel('Accuracy')

plt.show()
```



```
In [24]: """Fully Connected Model
Neurons in a fully connected layer have full connections to all activations in t
Adding another Dense Layer to model."""
def get_fc_model():
    model = Sequential([
        Lambda(standardize, input_shape=(28,28,1)),
        Flatten(),
        Dense(512, activation='relu'),
        Dense(10, activation='softmax')
    ])
    model.compile(optimizer='Adam', loss='categorical_crossentropy',
                  metrics=['accuracy'])
    return model
```

```
In [26]: fc = get_fc_model()
fc.optimizer.lr=0.01
```

```
In [27]: history = model.fit(
    batches,
    steps_per_epoch=len(batches),
    epochs=3,
    validation_data=val_batches,
    validation_steps=len(val_batches)
)
```

Epoch 1/3
591/591 ————— 2s 3ms/step - accuracy: 0.9264 - loss: 0.2636 - val_
 accuracy: 0.9181 - val_loss: 0.2951
 Epoch 2/3
591/591 ————— 2s 3ms/step - accuracy: 0.9285 - loss: 0.2595 - val_
 accuracy: 0.9107 - val_loss: 0.3089
 Epoch 3/3
591/591 ————— 2s 3ms/step - accuracy: 0.9281 - loss: 0.2572 - val_
 accuracy: 0.9133 - val_loss: 0.3027

```
In [28]: ## Convolutional Neural Network
from keras.layers import Convolution2D, MaxPooling2D

def get_cnn_model():
    model = Sequential([
        Lambda(standardize, input_shape=(28,28,1)),
        Convolution2D(32,(3,3), activation='relu'),
        Convolution2D(32,(3,3), activation='relu'),
        MaxPooling2D(),
        Convolution2D(64,(3,3), activation='relu'),
        Convolution2D(64,(3,3), activation='relu'),
        MaxPooling2D(),
        Flatten(),
        Dense(512, activation='relu'),
        Dense(10, activation='softmax')
    ])
    model.compile(Adam(), loss='categorical_crossentropy',
                  metrics=['accuracy'])
    return model
```

```
In [29]: model = get_cnn_model()
model.optimizer.lr=0.01
```

```
In [30]: history = model.fit(
    batches,
    steps_per_epoch=len(batches),
    epochs=3,
    validation_data=val_batches,
    validation_steps=len(val_batches)
)
```

Epoch 1/3
591/591 ————— 19s 28ms/step - accuracy: 0.9580 - loss: 0.1385 - va
 l_accuracy: 0.9771 - val_loss: 0.0722
 Epoch 2/3
591/591 ————— 17s 29ms/step - accuracy: 0.9868 - loss: 0.0417 - va
 l_accuracy: 0.9895 - val_loss: 0.0375
 Epoch 3/3
591/591 ————— 19s 32ms/step - accuracy: 0.9916 - loss: 0.0276 - va
 l_accuracy: 0.9881 - val_loss: 0.0432

```
In [31]: ## Data Augmentation
gen = ImageDataGenerator(rotation_range=8, width_shift_range=0.08, shear_range=0.
                        height_shift_range=0.08, zoom_range=0.08)
batches = gen.flow(X_train, y_train, batch_size=64)
val_batches = gen.flow(X_val, y_val, batch_size=64)
```

```
In [32]: model.optimizer.lr=0.001
history = model.fit(
    batches,
```

```

steps_per_epoch=len(batches),
epochs=3,
validation_data=val_batches,
validation_steps=len(val_batches)
)

```

Epoch 1/3

591/591 ————— 24s 41ms/step - accuracy: 0.9788 - loss: 0.0676 - val_accuracy: 0.9769 - val_loss: 0.0733

Epoch 2/3

591/591 ————— 23s 39ms/step - accuracy: 0.9868 - loss: 0.0423 - val_accuracy: 0.9860 - val_loss: 0.0429

Epoch 3/3

591/591 ————— 23s 39ms/step - accuracy: 0.9891 - loss: 0.0356 - val_accuracy: 0.9902 - val_loss: 0.0295

In [33]: `"""Adding Batch Normalization
it helps to fine tune hyperparameters more better and train really deep neural n
from tensorflow.keras.layers import BatchNormalization"""`

```

def get_bn_model():
    model = Sequential([
        Lambda(standardize, input_shape=(28,28,1)),
        Convolution2D(32,(3,3), activation='relu'),
        BatchNormalization(axis=1),
        Convolution2D(32,(3,3), activation='relu'),
        MaxPooling2D(),
        BatchNormalization(axis=1),
        Convolution2D(64,(3,3), activation='relu'),
        BatchNormalization(axis=1),
        Convolution2D(64,(3,3), activation='relu'),
        MaxPooling2D(),
        Flatten(),
        BatchNormalization(),
        Dense(512, activation='relu'),
        BatchNormalization(),
        Dense(10, activation='softmax')
    ])
    model.compile(Adam(), loss='categorical_crossentropy', metrics=['accuracy'])
    return model

```

In [34]: `model= get_bn_model()
model.optimizer.lr=0.01
history = model.fit(
 batches,
 steps_per_epoch=len(batches),
 epochs=3,
 validation_data=val_batches,
 validation_steps=len(val_batches)
)`

Epoch 1/3

591/591 ————— **32s** 51ms/step - accuracy: 0.9489 - loss: 0.1637 - val_accuracy: 0.9748 - val_loss: 0.0850

Epoch 2/3

591/591 ————— **33s** 56ms/step - accuracy: 0.9794 - loss: 0.0652 - val_accuracy: 0.9867 - val_loss: 0.0520

Epoch 3/3

591/591 ————— **35s** 58ms/step - accuracy: 0.9840 - loss: 0.0504 - val_accuracy: 0.9886 - val_loss: 0.0354

```
In [35]: model.optimizer.lr=0.01
          datagen = ImageDataGenerator()

          batches = gen.flow(X, y, batch_size=64)
          history = model.fit(
              batches,
              steps_per_epoch=len(batches),
              epochs=3,
              validation_data=val_batches,
              validation_steps=len(val_batches)
          )
```

Epoch 1/3

657/657 ————— **37s** 56ms/step - accuracy: 0.9859 - loss: 0.0471 - val_accuracy: 0.9907 - val_loss: 0.0297

Epoch 2/3

657/657 ————— **42s** 64ms/step - accuracy: 0.9878 - loss: 0.0392 - val_accuracy: 0.9886 - val_loss: 0.0342

Epoch 3/3

657/657 ————— **40s** 61ms/step - accuracy: 0.9878 - loss: 0.0379 - val_accuracy: 0.9914 - val_loss: 0.0246

```
In [37]: import numpy as np
          import pandas as pd

          # Get predictions (returns probabilities)
          predictions_proba = model.predict(X_test, verbose=0)

          # Convert to class labels (highest probability class)
          predictions = np.argmax(predictions_proba, axis=1)

          # Create submission DataFrame
          submissions = pd.DataFrame({
              "ImageId": list(range(1, len(predictions) + 1)),
              "Label": predictions
          })
          submissions.to_csv("output.csv", index=False, header=True)
```